

Perform EVAT GenerativeAI multi-table generation comparison

- Load the Original Data Model tables
- Define the original data model entity relationships
- Train the AI Model on the above data.
- Generate a new Dataset using AI synthesis.
- Compare the original data Model Versus the synthesized dataset/

Data preparation

Fetch & Review original Data Model Files

```
# prompt: load ChargingSession.csv into a dataframe ChargingSession, load OrderScheduling.csv into a dataframe OrderScheduling, load PurchaseWholesaleElec.csv into a dataframe PurchaseWholesaleElec, load Stations.csv into a dataframe Stations, load UserPayment.csv into a dataframe UserPayment, load Users.csv into a dataframe Users, load ev.csv into a dataframe ev
import pandas as pd

# Load the datasets into pandas DataFrames
ChargingSession = pd.read_csv('ChargingSession.csv')
OrderScheduling = pd.read_csv('OrderScheduling.csv')
PurchaseWholesaleElec = pd.read_csv('PurchaseWholesaleElec.csv')
Stations = pd.read_csv('Stations.csv')
UserPayment = pd.read_csv('UserPayment.csv')
Users = pd.read_csv('Users.csv')
ev = pd.read_csv('ev.csv')
```

ChargingSession.head(2)

	Id	SlotID	UserId	StationId	LocationId	ChargeStartTime	ChargeEndTime	ChargeTimeHrs	KwhTotal	CustPricePerKW	CustPricePerKwh
0	4926737	841344	78908148	730023	878393	7/10/2023 7:18	7/10/2023 8:44	0 days 01:26:00	82.049511	0.570110	0.570110
1	3738844	252642	78908148	730023	878393	5/08/2023 5:29	5/08/2023 7:01	0 days 01:32:00	88.246562	0.532488	0.532488

Next steps:

[Generate code with ChargingSession](#)

[View recommended plots](#)

[New interactive sheet](#)

OrderScheduling.head(2)

	Id	UserId	StationId	SlotId	ConnectorType	BookingDate	SlotStartTime	SlotLength	EstCost50KWCharger	BookingTimeS
0	4926737	78908148	730023	841344	CCS	7/10/2023	7	2	55.0	[7, 8]
1	3738844	78908148	730023	252642	CCS	5/08/2023	5	3	82.5	[5, 6, 7]

PurchaseWholesaleElec.head(2)

	Id	StationId	SupplyId	ExpenditureCode	RRPperKW	RRPperKW24hrs	BasePrice	PeakPriceAdjustment	CustPricePerKW	PurchasePricePerKwh
0	4926737	730023	VIC	Wholesale_cost	0.01156	0.020110	0.55	0	0.570110	7/10/2023
1	3738844	730023	VIC	Wholesale_cost	0.00895	-0.017512	0.55	0	0.532488	5/08/2023

Stations.head(2)

```
UserPayment.head(2)
```

```
Users.head(2)
```

```
ev.head(2)
```

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Train a multi-table generator

Configuring a mutli-table generator is simply done, by configuring each table, their primary key as well as all their foreign key relations.

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Configuring a mutli-table generator is simply done, by configuring each table, their primary key as well as all their foreign key relations.

```
!pip install -U mostlyai
```

```

Collecting mostlyai
  Downloading mostlyai-0.7.0-py3-none-any.whl.metadata (5.4 kB)
Collecting httpx<0.28.0,>=0.25.0 (from mostlyai)
  Downloading httpx-0.27.2-py3-none-any.whl.metadata (7.1 kB)
Requirement already satisfied: pandas<3.0.0,>=1.5.3 in /usr/local/lib/python3.10/dist-packages (from mostlyai) (2.2.2)
Requirement already satisfied: pyarrow>=14.0.0 in /usr/local/lib/python3.10/dist-packages (from mostlyai) (17.0.0)
Requirement already satisfied: pydantic<3.0.0,>=2.4.2 in /usr/local/lib/python3.10/dist-packages (from mostlyai) (2.10.4)
Requirement already satisfied: rich>=13.7.0 in /usr/local/lib/python3.10/dist-packages (from mostlyai) (13.9.4)
Requirement already satisfied: anyio in /usr/local/lib/python3.10/dist-packages (from httpx<0.28.0,>=0.25.0->mostlyai) (3.7.1)
Requirement already satisfied: certifi in /usr/local/lib/python3.10/dist-packages (from httpx<0.28.0,>=0.25.0->mostlyai) (2024.1


```

```
Requirement already satisfied: httpcore==1.* in /usr/local/lib/python3.10/dist-packages (from httpx<0.28.0,>=0.25.0->mostlyai) (
Requirement already satisfied: idna in /usr/local/lib/python3.10/dist-packages (from httpx<0.28.0,>=0.25.0->mostlyai) (3.10)
Requirement already satisfied: sniffio in /usr/local/lib/python3.10/dist-packages (from httpx<0.28.0,>=0.25.0->mostlyai) (1.3.1)
Requirement already satisfied: h11<0.15,>=0.13 in /usr/local/lib/python3.10/dist-packages (from httpcore==1.*->httpx<0.28.0,>=0
Requirement already satisfied: numpy>=1.22.4 in /usr/local/lib/python3.10/dist-packages (from pandas<3.0.0,>=1.5.3->mostlyai) (1
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.10/dist-packages (from pandas<3.0.0,>=1.5.3->mos
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas<3.0.0,>=1.5.3->mostlyai) (20
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.10/dist-packages (from pandas<3.0.0,>=1.5.3->mostlyai) (
Requirement already satisfied: annotated-types>=0.6.0 in /usr/local/lib/python3.10/dist-packages (from pydantic<3.0.0,>=2.4.2->n
Requirement already satisfied: pydantic-core==2.27.2 in /usr/local/lib/python3.10/dist-packages (from pydantic<3.0.0,>=2.4.2->mc
Requirement already satisfied: typing-extensions>=4.12.2 in /usr/local/lib/python3.10/dist-packages (from pydantic<3.0.0,>=2.4.2
Requirement already satisfied: markdown-it-py>=2.2.0 in /usr/local/lib/python3.10/dist-packages (from rich>=13.7.0->mostlyai) (3
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in /usr/local/lib/python3.10/dist-packages (from rich>=13.7.0->mostlyai)
Requirement already satisfied: mdurl~0.1 in /usr/local/lib/python3.10/dist-packages (from markdown-it-py>=2.2.0->rich>=13.7.0->
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.8.2->pandas<3.0.0,>
Requirement already satisfied: exceptiongroup in /usr/local/lib/python3.10/dist-packages (from anyio->httpx<0.28.0,>=0.25.0->mos
Downloading mostlyai-0.7.0-py3-none-any.whl (45 kB)
46.0/46.0 kB 2.0 MB/s eta 0:00:00
Downloading httpx-0.27.2-py3-none-any.whl (76 kB)
76.4/76.4 kB 4.1 MB/s eta 0:00:00
Installing collected packages: httpx, mostlyai
Attempting uninstall: httpx
Found existing installation: httpx 0.28.1
Uninstalling httpx-0.28.1:
Successfully uninstalled httpx-0.28.1
Successfully installed httpx-0.27.2 mostlyai-0.7.0
```

```
from mostlyai import MostlyAI
```

```
# initialize client
```

```
mostly = MostlyAI(api_key='mostly-a85530a98a3abb708825e71e667bb3df2350fbaa90eb7ba4fccbb84e0a07cf53')
```

 Connected to <https://app.mostly.ai> (v321) as john.collins@sensation360.com

```
# Define the configuration for the generator
```

```
config = {
    "name": "Multi-table - EVAT2",
    "tables": [
        {
            "name": "ev",
            "primary_key": "Id",
            "columns": [
                {"name": "Id", "type": "integer"},
                {"name": "Manufacturer", "type": "string"},
                {"name": "Model", "type": "string"},
                {"name": "VehicleYear", "type": "integer"},
                {"name": "VehicleClass", "type": "string"},
                {"name": "FuelType", "type": "string"},
                {"name": "ConnectorType", "type": "string"},
                {"name": "BatteryCapacityKwh", "type": "float"},
                {"name": "EstBattChargeTime80perc50KW", "type": "float"},
                {"name": "EnergyConsumptionWhkm", "type": "float"},
                {"name": "ElectricRangeKm", "type": "float"},
                {"name": "UserId", "type": "integer"}
            ],
            "data": ev,
        },
        {
            "name": "Users",
            "primary_key": "Id",
            "columns": [
                {"name": "Id", "type": "integer"},
                {"name": "UserFirstName", "type": "string"},
                {"name": "UserSurname", "type": "string"},
                {"name": "UserFullName", "type": "string"},
                {"name": "UserEmailAddress", "type": "string"},
                {"name": "UserPassword", "type": "string"},
                {"name": "UserRole", "type": "string"},
                {"name": "UserHomeAddress", "type": "string"},
                {"name": "UserMobilePhoneNumber", "type": "integer"},
                {"name": "UserAuthenticated", "type": "integer"},
                {"name": "VehicleID", "type": "integer"}
            ],
            "foreign_keys": [
                {"column": "VehicleID", "references": "ev.Id", "referencedTable": "ev", "is_context": True} # Added referencedTable
            ],
            "data": Users,
        },
    ],
}
```

```

{
  "name": "Stations",
  "primary_key": "Id",
  "columns": [
    {"name": "Id", "type": "integer"},
    {"name": "ChargingPoints", "type": "integer"},
    {"name": "Latitude", "type": "float"},
    {"name": "Longitude", "type": "float"},
    {"name": "OperatorID", "type": "string"},
    {"name": "Address", "type": "string"},
    {"name": "Amenities", "type": "string"},
    {"name": "BasePrice", "type": "float"},
    {"name": "PeakPriceAdjustment", "type": "float"},
    {"name": "EstCost50KWChargerHr", "type": "float"},
    {"name": "SlotIDs", "type": "string"},
    {"name": "ConnectionTypes", "type": "string"},
    {"name": "PaymentTypes", "type": "string"},
    {"name": "PowerOutput", "type": "float"},
    {"name": "LocationName", "type": "string"},
    {"name": "PostalCode", "type": "integer"},
    {"name": "LocationID", "type": "integer"}
  ],
  "data": Stations,
},
{
  "name": "OrderScheduling",
  "primary_key": "Id",
  "columns": [
    {"name": "Id", "type": "integer"},
    {"name": "UserId", "type": "integer"},
    {"name": "StationId", "type": "integer"},
    {"name": "SlotId", "type": "integer"},
    {"name": "ConnectorType", "type": "string"},
    {"name": "BookingDate", "type": "datetime"},
    {"name": "SlotStartTime", "type": "integer"},
    {"name": "SlotLength", "type": "integer"},
    {"name": "EstCost50KWCharger", "type": "float"},
    {"name": "BookingTimeSlots", "type": "string"},
    {"name": "BookingDeposit", "type": "float"}
  ],
  "foreign_keys": [
    {"column": "UserId", "references": "Users.Id", "referencedTable": "Users", "is_context": False}, # Added referenced
    {"column": "StationId", "references": "Stations.Id", "referencedTable": "Stations", "is_context": True} # Added refe
  ],
  "data": OrderScheduling
},
{
  "name": "UserPayment",
  "primary_key": "Id",
  "columns": [
    {"name": "Id", "type": "integer"},
    {"name": "BookingDeposit", "type": "float"},
    {"name": "PurchaseId", "type": "integer"},
    {"name": "PaymentType", "type": "string"},
    {"name": "CustPayAmount", "type": "float"},
    {"name": "TransAmount", "type": "float"},
    {"name": "UserId", "type": "integer"},
    {"name": "StationId", "type": "integer"},
    {"name": "PaymentDate", "type": "datetime"},
    {"name": "ChargingId", "type": "integer"},
    {"name": "OrderId", "type": "integer"}
  ],
  "foreign_keys": [
    {"column": "PurchaseId", "references": "PurchaseWholesaleElec.Id", "referencedTable": "PurchaseWholesaleElec", "is_c
    {"column": "StationId", "references": "Stations.Id", "referencedTable": "Stations", "is_context": False}, # Added re
    {"column": "UserId", "references": "Users.Id", "referencedTable": "Users", "is_context": False}, # Added referenced
    {"column": "OrderId", "references": "OrderScheduling.Id", "referencedTable": "OrderScheduling", "is_context": False}
    {"column": "ChargingId", "references": "ChargingSession.Id", "referencedTable": "ChargingSession", "is_context": Tru
  ],
  "data": UserPayment, # Use the UserPayment DataFrame directly
},
{
  "name": "PurchaseWholesaleElec",
  "primary_key": "Id",
  "columns": [
    {"name": "Id", "type": "integer"},
    {"name": "StationId", "type": "integer"},
    {"name": "SupplyId", "type": "string"},
    {"name": "ExpenditureCode", "type": "string"},
    {"name": "RRPperKW", "type": "float"},

```

```


        {"name": "RRPperKW24hrs", "type": "float"},
        {"name": "BasePrice", "type": "float"},
        {"name": "PeakPriceAdjustment", "type": "float"},
        {"name": "CustPricePerKW", "type": "float"},
        {"name": "PurchaseDate", "type": "datetime"}
    ],
    "foreign_keys": [
        {"column": "StationId", "references": "Stations.Id", "referencedTable": "Stations", "is_context": True} # Added refe
    ],
    "data": PurchaseWholesaleElec,
},
{
    "name": "ChargingSession",
    "primary_key": "Id",
    "columns": [
        {"name": "Id", "type": "integer"},
        {"name": "SlotID", "type": "integer"},
        {"name": "UserId", "type": "integer"},
        {"name": "StationId", "type": "integer"},
        {"name": "LocationId", "type": "integer"},
        {"name": "ChargeStartTime", "type": "datetime"},
        {"name": "ChargeEndTime", "type": "datetime"},
        {"name": "ChargeTimeHrs", "type": "string"},
        {"name": "KwhTotal", "type": "float"},
        {"name": "CustPricePerKW", "type": "float"},
        {"name": "CustPayAmount", "type": "float"},
        {"name": "BookingDate", "type": "datetime"},
        {"name": "ConnectorType", "type": "string"},
        {"name": "OrderId", "type": "integer"},
        {"name": "PurchaseId", "type": "integer"}
    ],
    "foreign_keys": [
        {"column": "OrderId", "references": "OrderScheduling.Id", "referencedTable": "OrderScheduling", "is_context": True},
        {"column": "UserId", "references": "Users.Id", "referencedTable": "Users", "is_context": False}, # Added referenced1
        {"column": "StationId", "references": "Stations.Id", "referencedTable": "Stations", "is_context": False}, # Added re
        {"column": "PurchaseId", "references": "PurchaseWholesaleElec.Id", "referencedTable": "PurchaseWholesaleElec", "is_c
    ],
    "data": ChargingSession
}
]
}
}

```

```

# Display the generator configuration
#config

```


 /usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transfo` and `should_run_async` (code)

```

# configure a generator, but don't yet start the training thereof
g = mostly.train(config=config, start=False)

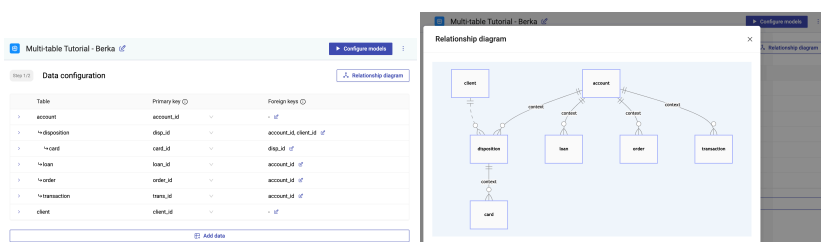
# open generator in a new browser tab
g.open()

```

 /usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transfo` and `should_run_async` (code)

Created generator [917375b2-3456-4feb-bd10-31af72ea82bf](https://app.mostly.ai/d/generators/917375b2-3456-4feb-bd10-31af72ea82bf)
 'https://app.mostly.ai/d/generators/917375b2-3456-4feb-bd10-31af72ea82bf'

You can now also inspect the configuration on the Web UI. It will look like this:



Now, launch the training, and wait for it to be finished. This shouldn't take longer than 10 minutes.

```

g.training.start()
g = g.training.wait(progress_bar=True)

```

```

/usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transf
and should_run_async(code)
Overall job progress _____ 100% 0:01:02
Step ev:tabular PULL_TRAINING_DATA _____ 100% 0:00:05
Step ev:tabular ANALYZE_TRAINING_DATA _____ 100% 0:00:05
Step ev:tabular ENCODE_TRAINING_DATA _____ 100% 0:00:02
Step ev:tabular TRAIN_MODEL _____ 100% 0:00:02
Step ev:tabular GENERATE_MODEL_REPORT_DATA _____ 100% 0:00:00
Step ev:tabular CREATE_MODEL_REPORT _____ 100% 0:00:00
Step Users:tabular PULL_TRAINING_DATA _____ 100% 0:00:02
Step Users:tabular ANALYZE_TRAINING_DATA _____ 100% 0:00:05
Step Users:tabular ENCODE_TRAINING_DATA _____ 100% 0:00:02
Step Users:tabular TRAIN_MODEL _____ 100% 0:00:02
Step Users:tabular GENERATE_MODEL_REPORT_DATA _____ 100% 0:00:00
Step Users:tabular CREATE_MODEL_REPORT _____ 100% 0:00:00
Step Stations:tabular PULL_TRAINING_DATA _____ 100% 0:00:05
Step Stations:tabular ANALYZE_TRAINING_DATA _____ 100% 0:00:02
Step Stations:tabular ENCODE_TRAINING_DATA _____ 100% 0:00:02
Step Stations:tabular TRAIN_MODEL _____ 100% 0:00:02
Step Stations:tabular GENERATE_MODEL_REPORT_DATA _____ 100% 0:00:00
Step Stations:tabular CREATE_MODEL_REPORT _____ 100% 0:00:10
Step OrderScheduling:tabular PULL_TRAINING_DATA _____ 100% 0:00:02
Step OrderScheduling:tabular ANALYZE_TRAINING_DATA _____ 100% 0:00:05
Step OrderScheduling:tabular ENCODE_TRAINING_DATA _____ 100% 0:00:02
Step OrderScheduling:tabular TRAIN_MODEL _____ 100% 0:00:05
Step OrderScheduling:tabular GENERATE_MODEL_REPORT_DATA _____ 100% 0:00:02
Step OrderScheduling:tabular CREATE_MODEL_REPORT _____ 100% 0:00:00
Step ChargingSession:tabular PULL_TRAINING_DATA _____ 100% 0:00:05
Step ChargingSession:tabular ANALYZE_TRAINING_DATA _____ 100% 0:00:05
Step ChargingSession:tabular ENCODE_TRAINING_DATA _____ 100% 0:00:05
Step ChargingSession:tabular TRAIN_MODEL _____ 100% 0:00:13
Step ChargingSession:tabular GENERATE_MODEL_REPORT_DATA _____ 100% 0:00:00
Step ChargingSession:tabular CREATE_MODEL_REPORT _____ 100% 0:00:13
Step UserPayment:tabular PULL_TRAINING_DATA _____ 100% 0:00:05
Step UserPayment:tabular ANALYZE_TRAINING_DATA _____ 100% 0:00:05
Step UserPayment:tabular ENCODE_TRAINING_DATA _____ 100% 0:00:05
Step UserPayment:tabular TRAIN_MODEL _____ 100% 0:00:13
Step UserPayment:tabular GENERATE_MODEL_REPORT_DATA _____ 100% 0:00:02
Step UserPayment:tabular CREATE_MODEL_REPORT _____ 100% 0:00:13
Step PurchaseWholesaleElec:tabular PULL_TRAINING_DATA _____ 100% 0:00:05
Step PurchaseWholesaleElec:tabular ANALYZE_TRAINING_DATA _____ 100% 0:00:05
Step PurchaseWholesaleElec:tabular ENCODE_TRAINING_DATA _____ 100% 0:00:02
Step PurchaseWholesaleElec:tabular TRAIN_MODEL _____ 100% 0:00:05
Step PurchaseWholesaleElec:tabular GENERATE_MODEL_REPORT_DATA _____ 100% 0:00:02

```

▼ Generate a multi-table dataset

Once, the training has completed, you can generate a multi-table dataset with. If you do not specify the sample size, then the platform will generate as many subject records, as there were in the original subject tables. Otherwise, you will need to specify for each subject table, the number of records, as these are independently sampled.

```

# use generator to create a synthetic dataset
sd = mostly.generate('917375b2-3456-4feb-bd10-31af72ea82bf', start=False)
sd.open()

```

```

/usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transf
and should_run_async(code)
Created synthetic dataset 66523b04-edb4-4c86-a504-483bbf548c4a with generator 92b02a05-3696-4667-b0aa-1ec3f575d61c
'https://app.mostly.ai/d/synthetic-datasets/66523b04-edb4-4c86-a504-483bbf548c4a'

```

```

sd.generation.start()
sd = sd.generation.wait(progress_bar=True)

```

```

/usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transf
and should_run_async(code)
Overall job progress 88% 0:02:02
Step Stations:tabular GENERATE_DATA 100% 0:00:00
Step Stations:tabular CREATE_DATA_REPORT 100% 0:00:07
Step OrderScheduling:tabular GENERATE_DATA 100% 0:00:07
Step OrderScheduling:tabular CREATE_DATA_REPORT 100% 0:00:00
Step PurchaseWholesaleElec:tabular GENERATE_DATA 100% 0:00:07
Step PurchaseWholesaleElec:tabular CREATE_DATA_REPORT 100% 0:00:00
Step ChargingSession:tabular GENERATE_DATA 100% 0:00:15
Step ChargingSession:tabular CREATE_DATA_REPORT 100% 0:00:10
Step UserPayment:tabular GENERATE_DATA 100% 0:00:25
Step UserPayment:tabular CREATE_DATA_REPORT 100% 0:00:07
Step ev:tabular GENERATE_DATA 100% 0:00:00
Step ev:tabular CREATE_DATA_REPORT 100% 0:00:00
Step Users:tabular GENERATE_DATA 100% 0:00:05
Step Users:tabular CREATE_DATA_REPORT 100% 0:00:00
Step common FINALIZE_GENERATION 0% 0:00:04
Step common DELIVER_DATA 0% -:-:-
Step FINALIZE_GENERATION failed

```

Fetch the Generated Data

```

# fetch configuration via API
sd = mostly.synthetic_datasets.get('54b2599b-fbc2-44f6-ba0f-d6707fa74ee3')
config = sd.config()
config

```



```

name': 'EVAT_SESSION',
'description': 'EVAT Charging Session Data',
'tables': [
  {
    'name': 'ev',
    'configuration': {
      'sample_size': 85,
      'sample_seed_connector_id': None,
      'sample_seed_dict': None,
      'sample_seed_data': None,
      'sampling_temperature': 1.0,
      'sampling_top_p': 1.0,
      'rebalancing': None,
      'imputation': None,
      'fairness': None,
      'tabular_compute': 'c5f0d5da-04d9-4099-8394-e1048a469a5a',
      'language_compute': None
    }
  },
  {
    'name': 'Users',
    'configuration': {
      'sample_size': None,
      'sample_seed_connector_id': None,
      'sample_seed_dict': None,
      'sample_seed_data': None,
      'sampling_temperature': 1.0,
      'sampling_top_p': 1.0,
      'rebalancing': None,
      'imputation': None,
      'fairness': None,
      'tabular_compute': 'c5f0d5da-04d9-4099-8394-e1048a469a5a',
      'language_compute': None
    }
  },
  {
    'name': 'Stations',
    'configuration': {
      'sample_size': 105,
      'sample_seed_connector_id': None,
      'sample_seed_dict': None,
      'sample_seed_data': None,
      'sampling_temperature': 1.0,
      'sampling_top_p': 1.0,
      'rebalancing': None,
      'imputation': None,
      'fairness': None,
      'tabular_compute': 'c5f0d5da-04d9-4099-8394-e1048a469a5a',
      'language_compute': 'd2dc8ce3-2861-4bea-95c2-99c01a2ed084'
    }
  },
  {
    'name': 'OrderScheduling',
    'configuration': {
      'sample_size': None,
      'sample_seed_connector_id': None,
      'sample_seed_dict': None,
      'sample_seed_data': None,
      'sampling_temperature': 1.0,
      'sampling_top_p': 1.0,
      'rebalancing': None,
      'imputation': None,
      'fairness': None,
      'tabular_compute': 'c5f0d5da-04d9-4099-8394-e1048a469a5a',
      'language_compute': None
    }
  },
  {
    'name': 'ChargingSession',
    'configuration': {
      'sample_size': None,
      'sample_seed_connector_id': None,
      'sample_seed_dict': None,
      'sample_seed_data': None,
      'sampling_temperature': 1.0,
      'sampling_top_p': 1.0,
      'rebalancing': None,
      'imputation': None,
      'fairness': None,
      'tabular_compute': 'c5f0d5da-04d9-4099-8394-e1048a469a5a',
      'language_compute': None
    }
  },
  {
    'name': 'UserPayment',
    'configuration': {

```



```


configuration : {
  'sample_size': None,
  'sample_seed_connector_id': None,
  'sample_seed_dict': None,
  'sample_seed_data': None,
  'sampling_temperature': 1.0,
  'sampling_top_p': 1.0,
  'rebalancing': None,
  'imputation': None,
  'fairness': None,
  'tabular_compute': 'c5f0d5da-04d9-4099-8394-e1048a469a5a',
  'language_compute': None
},
{
  'name': 'PurchaseWholesaleElec',
  'configuration': {
    'sample_size': None,
    'sample_seed_connector_id': None,
    'sample_seed_dict': None,
    'sample_seed_data': None,
    'sampling_temperature': 1.0,
    'sampling_top_p': 1.0,
    'rebalancing': None,
    'imputation': None,
    'fairness': None,
    'tabular_compute': 'c5f0d5da-04d9-4099-8394-e1048a469a5a',
    'language_compute': None
  }
}

```

```

# once done, fetch the synthetic data as dictionary of DataFrames
synthetics = sd.data()

```

 /usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transf` and `should_run_async(code)`

▼ Show sample records for each table

```

for k in synthetics:
  print("===", k, "===")
  display(synthetics[k].sample(n=3))

```

```

=== Users ===
/usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transf
and should_run_async(code)

  Id  UserFirstName  UserSurname  UserFullName  UserEmailAddress  UserPassword  UserRole  UserHomeAddress  UserMobil

31  mostlyff-aa94-  406b-862c-  17dd5d1332ca  _RARE_  _RARE_  _RARE_  _RARE_  _RARE_  user  _RARE_

28  mostly13-  a658-4b5a-  a0c5-  ef6a4bf964be  _RARE_  _RARE_  _RARE_  _RARE_  _RARE_  user  _RARE_

54  mostly57-  9bc7-4d41-  852e-  c0c8a82163cb  _RARE_  _RARE_  _RARE_  _RARE_  _RARE_  user  _RARE_

=== OrderScheduling ===

  Id  UserId  StationId  SlotId  ConnectorType  BookingDate  SlotStartTime  SlotLength  EstCost50KWCharger  Bool

3609  mostly68-  5cda-4802-  ac42-  147fd8084d2d  mostlyaa-  93d1-4ed2-  b515-  864050624e21  214539  CCS  2023-06-01  21  3  55.0

1902  mostly4f-  5877-4179-  a9a8-  e971da7ed7ff  mostlya2-  db7e-4114-  8569-  1a85f7b00ac7  567648  CCS  2023-05-01  15  2  55.0

3066  mostlyd9-  f5d0-49ab-  a526-  fac44398ebfc  mostly00-  2d92-49e6-  8d26-  c2da4cd41640  234781  CCS  NaT  17  2  55.0

=== Stations ===

  Id  ChargingPoints  Latitude  Longitude  OperatorID  Address

15  mostly20-  60a1-4576-  a0f0-  c0f3915b108b  3  -37.75771  144.955264  TBD

83  mostlyd1-  a726-4d9c-  a3b6-  d3b0079c8954  5  -37.802159  144.970827  TBD  144RmeTBDwen77intoobour.Amwerction  RdR37catio...  st50uitioceryoChargingPoints

61  mostly22-  17aa-4ea7-  8995-  26ec87db19bc  2  -37.783702  144.89731  TBD  E39StericUChargerHr66sPrPull  Restrooms WJ MelU...

=== UserPayment ===

  Id  BookingDeposit  PurchaseId  PaymentType  CustPayAmount  TransAmount  UserId  StationId  PaymentDate  Charg

1178  mostlya0-  bd1b-4800-  953b-  e35fc784ff73  1  7.0  Cash  27.957449  35.685663  7.0  0.0  NaT  mos: 09bb- ee7a0fb

1797  mostly28-f1e4-  4094-833a-  13b71269d386  1  94.0  MasterCard  32.84096  27.127117  6.0  14.0  NaT  mos: aee5- c588e61

1225  mostlyd4-  c6b0-44a1-  a5a0-  87c5f9937333  1  6.0  Cash  50.028035  30.306651  91.0  9.0  NaT  mos: 0c49- b54c5e8

=== ChargingSession ===

  Id  SlotID  UserId  StationId  LocationId  ChargeStartTime  ChargeEndTime  ChargeTimeHrs  KwhTotal  CustPricePer

2651  mostlyfb-ef95-  4d26-9343-  20d0754ea3ac  480509  -4454.0  1.0  181553  NaT  NaT  0 days 01:00:00  53.039271  0.582t

1064  mostlyf5-32d6-  499b-9eae-  0aa2c80c6048  176687  57.0  22.0  698346  2023-05-10  10:12:00  NaT  0 days 01:00:00  76.275731  0.692:

mostly35-

```


1/11/25, 9:49 PM

EVAT GenerativeAI multi-table generation comparison.ipynb - Colab

3250	mostly39- d334-43de- bfc5- 1111088c7afc	mostlydd- 170c-4b32- ba59- d18c3abb32c2	664572	13.0	19.0	492042	NaT	NaT	0 days 01:17:00	55.518466	0.532!
=== PurchaseWholesaleElec ===											
	Id	StationId	SupplyId	ExpenditureCode	RRPperKW	RRPperKW24hrs	BasePrice	PeakPriceAdjustment	CustPriceP		
822	mostly39- d334-43de- bfc5- 1111088c7afc	mostlydd- 170c-4b32- ba59- d18c3abb32c2	VIC	Wholesale_cost	0.22699	0.061748	0.55	0	0.57		
515	mostly43- 7b7a-42f0- a221- 2b29311a8dfc	mostlya2- 643e-4b71- b2a5- 649e3b150469	VIC	Wholesale_cost	-0.05388	-0.041298	0.55	0	0.68		
1563	mostlyd9- cc5e-4066- ad1f- 1749587ce654	mostly11- 3c2c-4b7c- a567- 7021089997cb	VIC	Wholesale_cost	0.127252	0.02317	0.55	0	0.69		
=== ev ===											
	Id	Manufacturer	Model	VehicleYear	VehicleClass	FuelType	ConnectorType	BatteryCapacityKwh	EstBattChargeTi		
69	mostly4f-242b- 49e9-b593- 1b6b2e803ac5	_RARE_	Tesla Model Y	2024	Medium SUV	BEV	CCS	94.977			
17	mostly10- e119-4a18- 84e9- 35de448044a6	BMW	_RARE_	2024	Small Car	BEV	CCS	88.524			
49	mostly2e-f3da- 4fa2-bbc3- 5626f934c0ae	BYD	MG MG4	2024	Small Car	BEV	CCS	81.487			

Comparison For Charging Session

original Charging table data
ChargingSession.head(2)




/usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transfrc` and `should_run_async` (code)


	Id	SlotID	UserId	StationId	LocationId	ChargeStartTime	ChargeEndTime	ChargeTimeHrs	KwhTotal	CustPricePerKW	Cu
0	4926737	841344	78908148	730023	878393	7/10/2023 7:18	7/10/2023 8:44	0 days 01:26:00	82.049511	0.570110	
1	3738844	252642	78908148	730023	878393	5/08/2023 5:29	5/08/2023 7:01	0 days 01:32:00	88.246562	0.532488	

Next steps:


Generate code with ChargingSession

 View recommended plots

New interactive sheet



 Generate

copy ChargingSession to dataframe Original_charging




Close

< 1 of 1 >

  [Use code with caution](#)

prompt: copy ChargingSession to dataframe Original_charging

Original_charging = ChargingSession.copy()



/usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transfrc` and `should_run_async` (code)

synthetics['ChargingSession'].head(2)



rnel.py:283: DeprecationWarning: `should_run_async` will not call `transfo

ChargeStartTime	ChargeEndTime	ChargeTimeHrs	KwhTotal	CustPricePerKW
NaT	NaT	_RARE_	70.019863	0.586903
2023-11-04 17:43:00	2023-07-04 18:46:00	_RARE_	47.591555	0.65694



Generate

copy synthetics['ChargingSession'] to dataframe Synthetics_charging

Q

Close

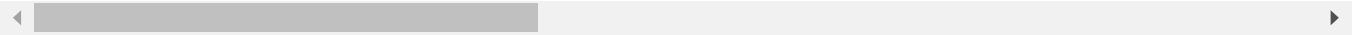
< 1 of 1 > [Use code with caution](#)

```
# prompt: copy synthetics['ChargingSession'] to dataframe Synthetics_charging

Synthetics_charging = synthetics['ChargingSession'].copy()
```



/usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transfo
and should_run_async(code)



Generate

Create a plot to compare the length of both dataframe Original_charging and Synthetics_charging

Q

Close

< 1 of 1 > [Use code with caution](#)


```
# prompt: Create a plot to compare the length of both dataframe Original_charging and Synthetics_charging

import matplotlib.pyplot as plt

# Assuming 'Original_charging' and 'Synthetics_charging' are already defined DataFrames.

# Calculate the lengths of the DataFrames.
original_length = len(Original_charging)
synthetic_length = len(Synthetics_charging)

# Create a bar plot to compare the lengths.
plt.figure(figsize=(8, 6))
plt.bar(['Original Charging', 'Synthetic Charging'], [original_length, synthetic_length], color=['skyblue', 'lightcoral'])
plt.ylabel('Number of Records')
plt.title('Comparison of DataFrame Lengths')
plt.show()
```

 /usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transf` and `should_run_async(code)`

Comparison of DataFrame Lengths

 Generate

Create a plot to compare the number of unique value in columns UserId, StationId, for both dataframe Original_charging and Synthetics_charging



Close

< 1 of 1 >   [Use code with caution](#)

prompt: Create a plot to compare the number of unique value in columns UserId, StationId, for both dataframe Original_charging and

```
import matplotlib.pyplot as plt

# Calculate the number of unique values for UserId and StationId in both DataFrames
unique_original_userid = Original_charging['UserId'].nunique()
unique_original_stationid = Original_charging['StationId'].nunique()

unique_synthetic_userid = Synthetics_charging['UserId'].nunique()
unique_synthetic_stationid = Synthetics_charging['StationId'].nunique()


# Create a bar plot for comparison
columns = ['UserId', 'StationId']
original_values = [unique_original_userid, unique_original_stationid]
synthetic_values = [unique_synthetic_userid, unique_synthetic_stationid]

x = range(len(columns)) # the label locations
width = 0.35 # the width of the bars

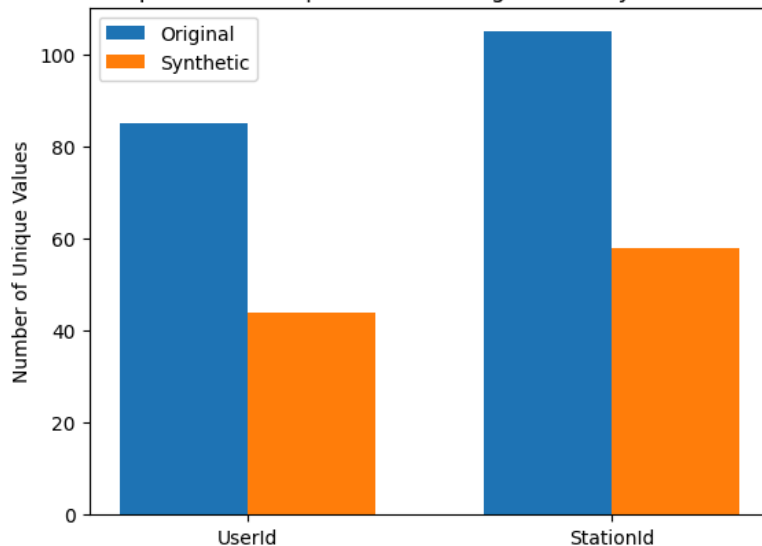
fig, ax = plt.subplots()
rects1 = ax.bar(x, original_values, width, label='Original')
rects2 = ax.bar([i + width for i in x], synthetic_values, width, label='Synthetic')

# Add some text for labels, title and custom x-axis tick labels, etc.
ax.set_ylabel('Number of Unique Values')
ax.set_title('Comparison of Unique Values in Original and Synthetic Data')
ax.set_xticks([i + width / 2 for i in x])
ax.set_xticklabels(columns)
ax.legend()

plt.show()
```

 /usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transf` and `should_run_async(code)`

Comparison of Unique Values in Original and Synthetic Data



 Generate

Plot the sns.hist Distribution of Energy Consumed kwhTotal for both dataframe Original_charging and Synthetics_charging



Close

< 1 of 1 >   [Use code with caution](#)

prompt: Plot the sns.hist Distribution of Energy Consumed kwhTotal for both dataframe Original_charging and Synthetics_charging


```
import seaborn as sns
import matplotlib.pyplot as plt

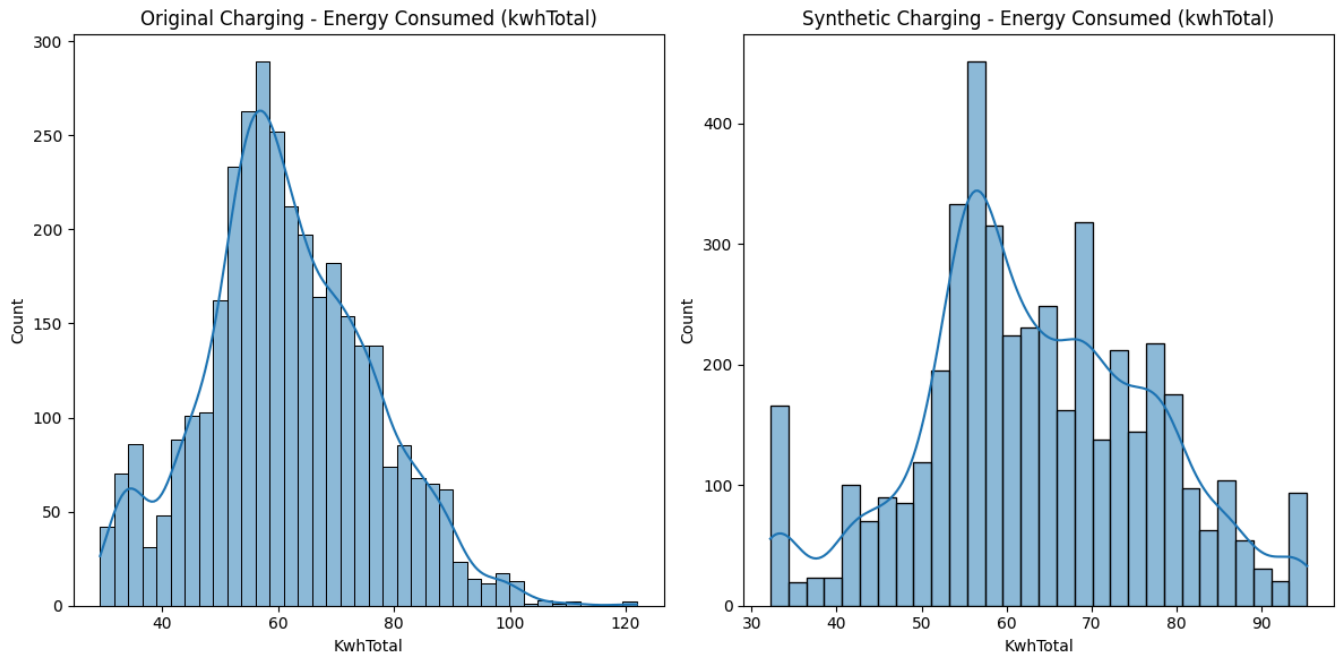
plt.figure(figsize=(12, 6))
```

```
plt.subplot(1, 2, 1)
sns.histplot(Original_charging['KwhTotal'], kde=True)
plt.title('Original Charging - Energy Consumed (kwhTotal)')

plt.subplot(1, 2, 2)
sns.histplot(Synthetics_charging['KwhTotal'], kde=True)
plt.title('Synthetic Charging - Energy Consumed (kwhTotal)')

plt.tight_layout()
plt.show()
```

 /usr/local/lib/python3.10/dist-packages/ipykernel/ipkernel.py:283: DeprecationWarning: `should_run_async` will not call `transf` and `should_run_async` (code)



 **Generate**  [Close](#)

< 1 of 1 >   [Use code with caution](#)

prompt: Plot the sns.hist Distribution of chargeTimeHrs for both dataframe Original_charging and Synthetics_charging

```
plt.figure(figsize=(12, 6))

plt.subplot(1, 2, 1)
sns.histplot(Original_charging['ChargeTimeHrs'], kde=True)
plt.title('Original Charging - Charge Time (hrs)')

plt.subplot(1, 2, 2)
sns.histplot(Synthetics_charging['ChargeTimeHrs'], kde=True)
plt.title('Synthetic Charging - Charge Time (hrs)')

plt.tight_layout()
plt.show()
```